

1 METHOD AND APPARATUS FOR PREDICTING MULTI-PART
2 PERFORMABILITY

3
4 BACKGROUND OF THE INVENTION

5
6 1. Field of the Invention:

7 The present invention relates generally to predicting a combined performance and
8 availability characteristic of complex systems -- the combination referred to hereinafter
9 as "performability". More specifically, the present invention relates to making efficient,
10 computerized predictions of the performability of systems composed of components that
11 can fail independently of one another. More particularly, the present invention includes a
12 method and apparatus for determining whether a predetermined multi-component system
13 can meet target performability requirements. An exemplary embodiment is described for
14 predicting multi-part performability of a predetermined, complex, multi-component, data
15 storage system.

16 2. Description of the Related Art:

17 Systems composed of many parts that can fail independently are commonplace.
18 Consider the simple example of a school bus system that uses a number of different buses
19 of different capacities; each bus can break down, independently of the others. Almost all
20 such systems have desired performance goals. In the example, it might be that all the
21 students be brought to school at the start of the school day in time for the first class.

22 It is natural to want such systems to be able to provide full service all the time,
23 but it is usually too expensive to ensure this. In such circumstances, people are usually
24 willing to live with a lower level of performance for a certain period, in order to reduce
25 the overall cost of the system. In the example, a school district might be able to afford a
26 spare bus or two -- but it would be unlikely to be able to keep a complete spare fleet of
27 buses. Even if there were plenty of spare buses, if a bus failed on its rounds, some of the
28 students might have to wait for a replacement bus to arrive, and so be late. It might even
29 be acceptable for no bus to be available for a day or two if alternative mechanisms
30 existed to get the students to school.

31 In general, in a system that can exhibit partial failures, the performance of the
32 system with one or more partial or complete component failures is often lower than when
33 the system is completely failure-free. The term "performability of a system" describes
34 the resulting set of performance levels achieved (perhaps under some load) and the
35 fractions of time that the system achieves them (as a result of partial or complete
36 component failures). The performability of a system can be represented as a set of pairs
37 " (r, f) ", where each " r " represents a performance level, and each " f " the fraction of time
38 that performance " r " is achieved or bettered. It is common to include pairs for both "full
39 performance" and "no performance" in the specification.

40 Performability can be predicted, or actually achieved (or measured) in practice.
41 "Performability requirements" are target goals for the performability of a system, and
42 "performability specifications" are written versions of performability requirements.
43 Similarly, a "performability function" is a representation of a performability requirement
44 (or measurement or prediction). It is sometimes shown as a "performability curve",
45 which is a graphical rendering of the function. Performability requirements (or
46 measurements or predictions) consist, by the previous definitions, of multiple parts; each

1 different combination of the performance and the availability represents a different part.
2 In what follows, we sometimes use the term “multi-part” without loss of generality, to
3 bring this fact to the attention of the reader.

4 A system “meets (or satisfies, or performs, or fulfills) its performability
5 requirements” (or “is predicted to . . .”) if the achieved (or predicted) performability of
6 the system is at least as good as the performability requirement for it. An object of this
7 invention is to provide a faster way to calculate whether this is the case.

8 In the example above, the ideal performance requirement is “all students get to
9 school on time, every school day”. But, given data on the rate at which school buses fail,
10 and the costs of having standby buses, the school district is likely to accept a lower level
11 of performance as long as the lower performance (fewer students on time) does not
12 happen too often and does not last too long. Thus, the requirements for the school bus
13 service for this example might read as follows (in part):

14 “Ideally, all 3000 students will get to school on time, every school day.

15 It is acceptable for up to 40 students to be delayed by 15 minutes, as long
16 as this does not happen more than 14 days a year.

17 It is acceptable for up to 20 students to be delayed by as much as an hour,
18 as long as this does not happen more than 7 days a year, provided that
19 each occurrence affects no more than 3 consecutive school days.

20 And, at most once a year, it is acceptable that schools close for one or two
21 days if too many buses all fail at the same time.”

22 These requirements combine performance (how many students get to school
23 when) with availability (the likelihood of a given number of students arriving on time).
24 A performability specification may contain or imply a “workload” as part of its
25 performance specifications (the number of students in this example), but not need do so.

26 The performability specification is a concise way to cope with the fact that the
27 average performance over a system’s lifetime is often not a useful metric for many target-
28 system designers. Even with all the data at hand, it is often difficult to work out whether
29 a particular system design is going to be able to meet a performability specification. For
30 example:

- 31 (a) the number of failure scenarios can be very large, and may have to
32 encompass multiple concurrent failures, not just one failed component at a
33 time; furthermore, each component may have different performance and
34 failure characteristics: so it may not be enough simply to predict the
35 effects of a single “representative” example failing;
- 36 (b) it may be expensive or difficult to predict the performance or likelihood of
37 each failure scenario; and
- 38 (c) the performability specifications may themselves be complicated.

39 Each of these issues will be discussed briefly below.

40 (a) *The number of failure scenarios can be very large.* In the example system,
41 assuming 100 buses in the fleet, there are 100 different cases of a single bus failing that
42 need to be analyzed. Each combination of possible failures – including the special case

1 of no failures – is a separate “failure scenario”. Analyzing 101 different failure scenarios
2 is not too hard using a computer. But if the school buses fail often enough that there is a
3 reasonable chance that two of them will be out of service, the number of combinations
4 increases dramatically – to approximately 100 times 100 cases, or about 10,000. Coping
5 with a third failed bus may involve almost 1 million failure scenarios to analyze. And
6 each bus may have its own, distinct failure mechanisms (for example, one might be have
7 a petrol engine, and another a diesel engine).

8 There are, of course, systems that are much larger still. For example, some
9 computer storage systems may contain thousands of components. The number of failure
10 scenarios grows exponentially as a function of the number of components in the system,
11 and of the number of concurrent failures it can tolerate while still remaining “functional.”
12 For example, if a computer storage system with 10,000 data storage disks could tolerate a
13 maximum of two disk failures, then it would be necessary to evaluate its performance in
14 each of 100,000,000 different failure scenarios.

15 (b) *It may be expensive or difficult to predict the performance or likelihood of*
16 *each failure scenario.* A “target system” being designed may be very complicated, the
17 workloads may be very complicated, the components may themselves be complex
18 systems, the models used to predict performance may be very complicated, slow, or
19 difficult to use, or some combination of these, may all be relevant to performability.
20 Analysis is even more difficult if different components have different failure rates or
21 performance characteristics. Returning to the simple example system, a fleet may
22 consist of different kinds of school buses, each with its own failure rate, speed, and
23 capacity.

24 (c) *The performability specifications may themselves be complicated.* In the
25 school bus example above, the number of students was constant; but it could also be the
26 case that the number varies; perhaps the school population fluctuates at different times of
27 year, or perhaps the number of students trying to get to school drops in very cold weather,
28 at precisely the time when buses are more likely to break. Such complications make the
29 problem of determining whether a given design (in this case, the number and type of
30 buses) will meet its performability specification even more difficult.

31 In computer storage systems, the workload may be very complicated indeed: it is
32 necessary to describe dozens of pieces of information about each portion of the workload
33 to be able to predict accurately its performance.

34 The number of possible combinations of different types of workload is effectively
35 extremely large.

36 Therefore, there is a need for a computerized system that can determine,
37 economically and efficiently, whether a given complex system design meets a given
38 performability specification.

39 3. **Problem Statement:** (applied to an exemplary data storage system).

40 This invention is especially suited for assisting with the design of multi-
41 component computer-based systems (for brevity, generally referred to hereinafter merely
42 as “systems” in the context of this document). Such systems typically comprise a
43 collection of one or more data processing, communication, and storage components. The
44 design, configuration, and maintenance of such computer systems is difficult. There is a
45 need to predict the behavior of the systems in a novel manner such that many system

1 design problems will be alleviated, including problems such as: allowing systems that
2 were installed to meet their requirements more often; reducing the cost of systems,
3 because there would be less need for expensive over-design; and reducing the number of
4 emergency repair and preemptive maintenance calls by having a good understanding
5 about which particular failures are relatively benign.

6 Generally, and as will be described in more depth hereinafter with respect to an
7 exemplary embodiment, complex systems, such as those comprising hundreds and
8 thousands of computers and attendant mass data storage apparatus, create complex
9 problems regarding selection, configuration, and maintenance for system designers and
10 information technology system administrators. The purchase and maintenance of such
11 systems are expensive costs of doing business. Pre-purchase configuring of such systems
12 and modeling performance without building prototypes is a complex task due to the
13 nearly infinite number of possible workloads and system configurations.

14 As shown in Figure 1, a representative system used hereinafter as an exemplary
15 embodiment for discussion of the present invention, the target system 100 is a relatively
16 small computer system having a disk array 132. We describe the invention in the context
17 of a performability analysis of the data storage subsystem. No limitation on the scope of
18 the invention is intended by the inventors nor should any be implied from the use of this
19 example.

20 As an example of how the system 100 of Figure 1 may be used, consider a health
21 maintenance organization (HMO) where host A 111 processes patient records data while
22 host B 111' processes provider and employer/employee data. Storage requirements for
23 profiling thousands of patients records and provider records and employer/employee
24 information may require a storage capacity of hundreds of gigabytes (GB), or even more
25 than a terabyte (TB), for the associated data. Based on the currently available hardware
26 technology, such a system 100 for the HMO might require seventy disk drives and ten
27 controllers.

28 The data storage subsystem of the system 100 includes the disk array 132. The
29 disk array 132 typically contains several disk drives 101-108 to store data, one or more
30 controllers 121, 121' both to communicate with the clients (host computers 111, 111')
31 and to control the operation of the disk array, one or more data caches 131, 131' to hold
32 stored data temporarily in a manner that is very fast to access, thereby improving
33 performance, and appropriate busses 109 or other interconnections to join these
34 components together. Other, associated components such as device drivers, disk array
35 firmware, and modules that implement basic functions in the data path (e.g., parity
36 calculation engines, direct memory access engines, busses, bus bridges, communication
37 adapters for busses and external networks, and the like, not shown) are also part of a
38 typical data storage subsystem.

39 There are many different designs of disk arrays, but most of them share one
40 common trait: they are intended to increase the likelihood that data stored in them will
41 survive the failure of one or more of the disk array's components. Because of this, such
42 disk arrays are often used for large data storage and management applications, often
43 storing very critical data.

44 The basic approach to achieving acceptable partial failure modes of operation is to
45 provide redundancy – the provision of more components than strictly needed to
46 accomplish the basic function. A redundant array of inexpensive disks ("RAID") is often

1 used for large data storage and management applications (slower, high capacity, tape
2 drive arrays and optical disk drive apparatus can be employed similarly). RAID was first
3 popularized by researchers at the University of California, Berkeley: see e.g., D.
4 Patterson, G. Gibson and R. Katz, *A Case for Redundant Arrays of Inexpensive Disks*
5 (*RAID*), Proceedings of the 1988 SIGMOD International Conference on the Management
6 of Data, Chicago, Illinois, May 1988. These RAID ideas were initially only applied to
7 the disk drives in a disk array, whereas, it is important to realize that true failure scenarios
8 affect the other data storage subsystem components (namely, all of Figure 1 excluding the
9 Host A 111 and Host B 111' computers). In one mode of RAID operation, multiple
10 copies of the stored data are kept, each copy on a different disk drive. This is often
11 referred to as "RAID 1 mode", or "mirroring". Although this increases the number of
12 disk drives needed, it also increases the availability of the data; if one disk drive breaks,
13 another is there to provide a copy of the data. Other RAID modes allow for partial
14 redundancy; these provide failure tolerance at lower cost, but with a more significant
15 performance degradation after a failure and during certain normal-mode input/output
16 ("IO") operations.

17 Data inside the array is spread out onto what is referred to in the art as Logical
18 Units ("LUs"), sometimes referred to as "LUNs" for Logical Unit Numbers, which are
19 the names by which LUs are known. Each LU represents a subset of the total storage
20 space available in the array, and is an aggregation of disk drives into a single logical
21 construct, visible to the host computers that use the array. Each LU is used and managed
22 independently of the other LUs. Typically, LUs can be constructed from any aggregation
23 of the disk drives accessible to a controller inside an array. In the Figure 1, four disk
24 drives 101, 102, 103, 104 are grouped into one such LU, and the remaining disk drives
25 105, 106, 107, 108 into another. (It is also common in the art to say that the LU is
26 "placed on" the disk drives.) Data flow is indicated generally by phantom lines,
27 demonstrating that in the example system shown in the Figure 1, host computer A 111
28 uses the LU on disk drives 101, 102, 103, 104, and host computer B 111' uses the LU on
29 disk drives 105, 106, 107, 108.

30 The introduction of redundancy in disk arrays – that is primarily present to handle
31 component failures – raises several issues about performance, including:

32 (a) although it may be possible to continue operation after a failure, the
33 performance of the system usually degrades in this state;

34 (b) even in failure-free mode, data replication can improve performance (e.g., if
35 two disks contain identical copies of the same data, it is possible to read it from the disk
36 that is less busy to improve performance), or hurt it (e.g., with two copies, an update has
37 to be sent to both before the operation is completed, and that takes longer than simply
38 updating one copy); and

39 (c) when failures are being repaired, recovery may impact performance: for
40 example, after a disk drive has failed, the system will begin rebuilding its contents on a
41 spare disk drive, so accesses originated by the reconstruction task compete with normal
42 accesses.

43 The workloads processed by computer storage systems are often themselves very
44 complicated, requiring a great many parameters to characterize them accurately. Small
45 variations in such workloads can sometimes have large effects on the performance and
46 cost of the associated storage systems. In the system 100 of Figure 1, a workload can be

1 defined by the stream of IO requests (READ and WRITE commands) issued by hosts A
2 111 and B 111' to the disk array. The workload can be characterized by parameters such
3 as the typical IO request size, and the rate at which such requests are generated. But this
4 is not enough to predict the performance of the storage subsystem of Figure 1 accurately.
5 To do so involves including many other workload characteristics, such as the amount of
6 the workload that can be cached in cache memories 131 and 131', the degree of
7 sequential accesses to the on-disk data, and in some circumstances, correlations in the
8 access patterns between the two hosts 111, 111'. Each of these can have a significant
9 impact on the performance and cost of the storage system needed to meet the needs of the
10 workload.

11 Developing designs for such systems is hard enough when failures are not taken
12 into account. There are many published papers on estimating the performance
13 characteristics of disk arrays. One example is the paper by Lee and Katz on *An Analytic
14 Performance Model of Disk Arrays*, published in the proceedings of the ACM
15 SIGMETRICS conference, May 1993 (pages 98—109).

16 Since a primary reason for deploying disk arrays is to support continued operation
17 in the presence of failures, it is often very important to be able to predict the
18 performability of the system under the range of failures that the system is likely to
19 encounter in practice. Doing so with manual techniques is slow, error-prone, and
20 unlikely to be satisfactory. It also suffers from all of the problems outlined above.

21 4. Description of the Prior Art:

22 Given the complexity of designing and managing large computer systems, system
23 designers rely on very simple rules of thumb to make design, configuration and purchase
24 decisions. This can lead to systems that do not satisfy their performance expectations, or
25 to excessive system cost due to over-design, or both.

26 A common approach is to build an actual, test system for a proposed workload in
27 order to evaluate system performance empirically. Although this is often time
28 consuming, hard to do, and expensive, it is nonetheless often the method employed by a
29 system supplier to demonstrate the viability of a proposed system to a potential customer.
30 Because of the difficulty of working with such systems, it is rare to explore even simple
31 failure modes because doing so is expensive, time consuming, and may not even be
32 possible.

33 Another technique in the prior art is to predict the performance of storage devices
34 with a performance model, such as that described in the paper by Lee and Katz on *An
35 Analytic Performance Model of Disk Arrays*. This approach is not a complete multi-part
36 performability analysis as defined herein because it fails to take into account the multi-
37 part performability requirements that are the essence of this invention. It is hard to
38 predict the performance of a single configuration using these tools, let alone explore
39 hundreds or thousands of failure scenarios.

40 Moreover, average performance over the system's lifetime is not a useful metric
41 for many target-system designers.

42 There is a need for a more realistic metric for the suitability of a given system for
43 a given task. Such a metric comes from putting together the concepts of availability and
44 performance, during both failure-free and degraded modes of operation under a given
45 workload.

Prior solutions generally treat the failure analysis of complex systems as Markov chain reward models or Petri net models. These methods work best when predicting whether a system meets a single-part performability specification, for example, the average bandwidth available in a disk array, taking into account the possibility of disk component failures. See e.g., S.M. Rezaul Islam, *Performability Analysis of Disk Arrays*, Proceedings of 36th Midwest Symposium on Circuits and Systems, Detroit, MI, August 16-18 1993, pp. 158-160, IEEE Publications; Ing-Ray Chen, *Effect of Probabilistic Error Checking Procedures on Performability of Robust Objects*, Proceedings of 8th ACM/SIGAPP Symposium on Applied Computing, Indianapolis, IN, February 14-16 1993, pp. 677-681, ACM Publications. Both of these papers use a different definition of performability than we do. They represent the system as a Markov chain and associate a single-valued “reward” with each state in the Markov chain. They compute the expected value of this reward based on all the state probabilities. Our definition of performability is multi-part; it does not require that the underlying system be modeled as a Markov chain, and we do not require all state probabilities to be calculated.

None of the above approaches provides an efficient, effective, high-quality prediction of whether a target system will meet its performability goals.

Thus, given a description of a complex system, and the performability requirements of the end-user or target-system designer, there is a need for a method and apparatus to provide a multi-part performability assessment as to whether the candidate system meets the requirements.

SUMMARY OF THE INVENTION

In general, the present invention provides a method and apparatus for calculating whether or not a target system will be able to meet its performance goals. The method uses a failure-scenario generator, a performance predictor, and a performance evaluator to perform this analysis.

In its basic aspects, the present invention provides a method of determining a multi-component target system performance capability with respect to a set of multi-part performability requirements, the method including: operating on a representation of the target system, providing a first failure-scenario analysis of said target system; generating a multi-part performability function of said target system using said first failure-scenario analysis; comparing said multi-part performability function with said set of multi-part performability requirements; and determining from said comparing whether said target system meets said multi-part performability requirements.

In another aspect, the present invention provides a computer memory including: computer code operating on a representation of the target system, providing a first failure-scenario analysis of said target system; computer code generating a multi-part performanceability function using said first failure-scenario analysis; computer code comparing said multi-part performanceability function with said multi-part performance requirements; and computer code determining from said comparing whether said target system has a capability of performing said multi-part performanceability requirements.

In another aspect, the present invention provides a method of doing business of verifying performability of a target system having predetermined components and predetermined multi-part performability requirements, the method including: using a computer, (1) operating on a representation of the target system, including providing a failure-scenario analysis of said target system; (2) generating a multi-part performability curve using said failure-scenario analysis; (3) comparing said requirements with said curve; (4) determining from said comparing whether said target system has the capability of performing said multi-part performability requirements; and (5) generating a report indicating of results whether said target system has the capability of performing said multi-part performability requirements.

The foregoing summary is not intended by the inventors to be an inclusive list of all the aspects, objects, advantages, or features of the present invention nor should any limitation on the scope of the invention be implied therefrom. This Summary is provided in accordance with the mandate of 37 C.F.R. 1.73 and M.P.E.P. 608.01(d) merely to apprise the public, and more especially those interested in the particular art to which the invention relates, of the basic nature of the invention in order to be of assistance in aiding ready understanding of the patent in future searches. Specific aspects, objects, advantages, and features of the present invention will become apparent upon consideration of the following explanation and the accompanying drawings, in which like reference designations represent like features throughout the drawings.

1 BRIEF DESCRIPTION OF THE DRAWINGS

2 Figure 1 is a schematic block diagram that depicts an exemplary computer system
3 including a disk array for its data storage subsystem;

4 Figure 2 is an exemplary graph depicting target-system designer predetermined
5 performability requirements;

6 Figure 3 is a graph depicting target-system designer predetermined performability
7 requirements as shown in Figure 2 and multi-part performability assessment results as
8 determined in accordance with the present invention for an exemplary system in which
9 the system satisfies the requirements;

10 Figure 3A is a graph combining depicting target-system designer predetermined
11 performability requirements as shown in Figure 2 and multi-part performability
12 assessment results as determined in accordance with the present invention for an
13 exemplary system that does not satisfy the requirements;

14 Figure 4 is a graph representing target-system designer predetermined
15 performability as shown in Figure 3, for an exemplary system where multi-part
16 performability does not need to be evaluated in particular operational regions, depicting a
17 particular result of a multi-part performability assessment;

18 Figure 5 is a block diagram of an architecture to evaluate multi-part
19 performability in accordance with the present invention; and

20 Figure 6 is a flow chart illustrating the methodology performed in accordance
21 with the present invention.

TECHNICAL FIELD

DESCRIPTION OF THE PRESENT INVENTION

Reference is made now in detail to a specific embodiment of the present invention that illustrates the best mode presently contemplated by the inventors for practicing the invention. Alternative embodiments are also briefly described as applicable. The sub-headings provided in this document are merely for the convenience of the reader; no limitation on the scope of the invention is intended nor should any be implied therefrom. Also, incorporated by reference, in entirety, is a twenty-one page appendix, Lee and Katz, *An Analytic Performance Model of Disk Arrays*, copyright 1993, ACM 0-89791-581-X/93/0005/0098, pp. 98-109.

For the purpose of this invention, whatever the nature of the actual system being analyzed, the performability requirements must be definable in a way that is comprehensive enough to capture all relevant parameters and is computer-readable. It is desirable – but not necessary – for these descriptions to be compact and understandable enough to be manipulated by both end users and target-system designers. What follows is an English-language version of a sample performability description for the previous example system 100:

For LU #1, the best-case performance of the system shall at least meet the following specification: a capacity of 5 Gigabytes; an average request rate of 100 IO/s (IO operations per second); a mean IO request size of 32 Kilobytes; and 70% of the accesses will be to sequential addresses.

During times of partial failure, the system shall support at least 50 IO/s for at least 40% of the time, and it shall support at least 10 IO/s for at least another 40% of the time. During these times of degraded performance, the remainder of the performance specification (capacity, request size, sequentiality) remains the same as above.

The system shall be completely non-functional for no more than 20% of the time.

Thus, a specific multi-part performability requirement is now defined as an expressible set of end-user or target-system designer-defined minimal criteria that the system under study must satisfy, where the expression is adapted to a computerized process for predicting performability of that system.

The processes presented with respect to this invention also require a computer-readable system description that contains information about the components of the system, their interconnections, and their performance and dependability characteristics (e.g. bandwidth, capacity, and failure modes and failure mode and repair rates).

Basic Operation:

In the main, the present invention provides a method and apparatus for determining whether a target system configuration (e.g., as exemplified in Figure 1) satisfies a given system performability requirement. That is, the system must provide given minimum levels of performance during given fractions of its operational lifetime.

The invention operates by generating failure scenarios for the system configuration and predicting the system's performance for each of them. Each successive prediction is incorporated into a set of system states that have been so evaluated. As soon

1 as a decision can be made about whether the system satisfies the requirements, or it does
2 not, the multi-part performability assessment process is halted and the assessment results
3 reported. The present invention may be implemented in software, or firmware, or a
4 combination of both.

5 In order to describe the present invention, an exemplary embodiment employing
6 graphical analysis is presented. It will be recognized by those skilled in the art that this
7 graphical representation is merely one form of analysis and that others may be employed
8 within the scope of the invention as defined in the claims; no limitation on the scope of
9 the invention is intended nor should any be implied.
10

11 Figure 2 is a graph exemplifying a very simple, predefined, multi-part
12 performability requirement. The target-system designer has specified, as a system
13 function here represented by “requirements curve” 201, that the system be available 40%
14 of the time at a predetermined performance level of at least 50 IO/s, represented as the
15 hatched area 203 below the requirements curve 201, and at least at 10 IO/s for 40% more
16 of the time, represented as the banded area 205 below the curve 201, and can be
17 completely down no more than 20% of the time, region 207. This equates to allowing
18 less than 50 IO/s for less than 60% of the time, region 209, and less than 10 IO/s for less
19 than 20% of the time, region 207. It should be recognized by those skilled in the art that
20 this exemplary requirements curve 201 is in essence any function representative of a set
21 of performance levels versus percentage of time at each of said performance levels.
22

23 More precisely, we denote a system’s condition regarding component failures by
24 its state, denoted S_i ; each state represents a particular combination of components that
25 have failed. We refer to the set “S” that contains all “M” possible system states as:
26

$$S = \{S_1, S_2, S_3, \dots, S_M\} \quad (\text{Equation 1}),$$

27
28 “M” can be a very large number. Associated with each state “ S_i ,” where “i” = 1 through
29 “M”, in the set “S” is a probability “ $OP(S_i)$ ” of the system being in that state, and a
30 performance “ $U(S_i)$ ” that the system demonstrates while in that state “ S_i ”. A “failure
31 scenario” sc_i is a set of one or more system states “ S_i' ” that have similar performance. In
32 the present invention no assumptions are made about the performance metric “ $U(S_i)$ "; it
33 is only assumed that given two different levels of performance, “a” and “b,” it is possible
34 to determine whether level “b” is as good as, or better than, level “a” (we denote this
35 condition as “ $b \geq a$ ”). The present invention will thus achieve a comparison condition for
36 the performance levels if each one of them is a scalar value.

37 In the example above, we used the single scalar value of “IO/s” as a simple
38 example of one employable performance metric. It is also possible to implement the
39 invention with richer forms of performance metrics. For example, if the composite
40 performance metric contains two values (say bandwidth “w” and latency “l”), then a
41 comparison performance metric can be straightforwardly defined as the pair (w,l) , where
42 $(w,l) \geq (w',l')$ if and only if $w \geq w'$ and $l \leq l'$. Any arbitrary number of scalar metrics can
43 be composed in this way. It will also be apparent to one skilled in the art that there are a
44 great many other possible ways of doing this comparison between multiple scalar
45 metrics.
46

Performability specifications are provided as a set of pairs, each pair comprising:

(1) a performance requirement (for the example system shown in Figure 1, this might be measured in number of IO requests per second that complete with less than a given response time);

(2) the fraction of the time the associated performance level is acceptable.

For example, a performability specification might require 50 IO/s or more at least 40% of the time; and 10 IO/s or more for at least 80% of the time.

More formally, the target-system designer provides the multi-part performability requirement, defined by "n" pairs of numbers $(r_1, f_1), (r_2, f_2), \dots, (r_n, f_n)$, where "n" is any positive integer, with the properties

$r_1 > r_2 > r_3 > \dots > r_n$, and
 $f_1 < f_2 < f_3 < \dots < f_n < 1$,

where each “r” denotes a performance level (e.g., 50 IO/s), and the corresponding “f” denotes the fraction of the system’s lifetime during which it delivers performance “r” or better.

A system is said to fulfill the target-system designer's multi-part performance requirement if for each performance level " r_i ", where $i=1,\dots,n$, there are sufficient number of possible system states where the system delivers performance level " r_i " or better, and they occur for a sufficient portion of the time. In other words, we need to determine whether the sum of the probabilities of all states that deliver performance level " r_i " or better is greater than or equal to " f_i ", the probability of being in a state where the system needs to deliver performance level r_i or better in the specified performanceability requirement.

More formally, a system can fulfill the target-system designer's multi-part performability requirement if:

$$\sum_{i=1}^M \text{OP}(S_i) \cdot 1(U(S_i) \geq r_j) \geq f_j, \text{ for } j=1,\dots,n \quad (\text{Equation 2})$$

The indicator function “ $1(e)$ ” in Equation 2 is equal to one if the expression “ e ” is true and zero otherwise.

The goal of this invention is to efficiently determine whether Equation 2 holds.

Multi-Part Performability - General:

The detailed computing of a multi-part performability assessment will now be described. It will use as an example the multi-part performability requirements illustrated in Figure 2 and a very simple candidate system that can be in one of only three failure scenarios during its operation as follows: (1) system down, 10% of the time, (2) at least 70 IO/s 60% of the time, and (3) fully functional at 100 IO/s 30% of the time.

Turning to Figure 3, in accordance with the exemplary embodiment of the present invention, a “multi-part performability curve” 301 is computed and plotted in comparison against the “multi-part performance requirement curve” 201. Note that at all times, for this example, the multi-part performance curve 301 is above the requirements curve

1 201. Therefore, the proposed system satisfies the target-system designer's multi-part
2 performability requirements. Any cross-over between the multi-part performability curve
3 301 and requirements curve 201 indicates a failure to meet the need with the specified
4 system; see for comparison the example in Figure 3A.

5 For a complex system, however, calculating all the possible states is virtually
6 impossible or at least very expensive. A system having a thousand independent
7 components, each of which can be digitally described as either "WORKING" ("1") or
8 "FAILED" ("0"), has in principle 2^{1000} failure-scenarios. Using a performance evaluator
9 on each of the millions of millions of states is impracticable with state of the art
10 computing systems. Instead, various optimizations are applied in the present invention to
11 reduce the number of states to an acceptable level for analysis. To achieve this, the
12 present invention takes advantage of existing information. For example, the
13 manufacturer of a disk array may provide specifications that indicate that the disk array
14 will be in a "failed" state no more than 10% of the time, as shown in Figure 4 with
15 shaded area 401. Similarly, the manufacturer may provide guarantees about the portion
16 of the time that the system is fully operational at 100 IO/s (Figure 4, shaded area 402).
17 The failure scenario generator 502 may take advantage of this to generate as the first two
18 scenarios a "fully operational" scenario and a "failed" scenario, each with the
19 corresponding probability specified by the manufacturer. The failed scenario may cover
20 a number of states, each corresponding to a set of component failures that render the
21 device unusable. Similarly, the fully operational scenario might cover a number of states
22 in which the component failures do not affect the device performance (for example, in a
23 device with dual redundant power supplies, the device might deliver full performance
24 even with one power supply failed). Using these pre-aggregated, pre-defined, failure
25 scenarios reduces the number of states remaining to be considered and therefore the time
26 required for the analysis. Note that if this existing information is not available, the
27 method described here still works, but may need to analyze more failure scenarios.
28

29 *Multi-Part Performability Evaluator:*

30 Figure 5 is a block diagram of multi-part performability assessment, or
31 evaluation, architecture 500 in accordance with the present invention. Three subsystems
32 (computer code routines or modules) are provided:

- 33 a performability evaluator 501,
34 a failure-scenario generator 502, and
35 a performance predictor 503.
36

37 *Failure-Scenario Generator 502:*

38 The failure-scenario generator 502 takes as input, represented by an arrow labeled
39 [3], the description of a candidate system and generates pairs (sc_i, p_i) (represented by the
40 arrow labeled [4]). Each " sc_i " is a failure scenario, and corresponds to one or more
41 system states " S_i ". Each state corresponds to the system with zero or more components
42 marked as failed. Next, " p_i " is the estimated fraction of time during that the input system
43 will be in one of the states in failure scenario " sc_i ".

44 Many computation methods are known in the art for calculating the probability of
45 a particular state. A variety of implementations can be employed in accordance with the

1 present invention. In the main, a failure-scenario generator is an algorithm generating a
2 set of successive failure scenarios: $sc_1, sc_2, \dots, sc_i, \dots, sc_n$, where sc_i = current scenario.

3 Each failure scenario consists of one or more failure states, each failure state
4 being representative of a failure mode in which one or more critical components of the
5 system has failed. Each failure scenario “ sc_i ” has an occurrence probability “ p_i ”, that is
6 computed by the failure-scenario generator.

7 Failure states can be grouped into failure scenarios in a variety of ways. In
8 accordance with the preferred embodiment of the present invention, failure scenarios are
9 generated in approximate order of decreasing likelihood (i.e. the multi-part performability
10 assessment will start with most likely failure scenarios). This avoids unnecessary
11 computations by concentrating the assessment efforts on the states that carry more weight
12 in the multi-part performability calculation. Each failure scenario can then be analyzed
13 more intensively to determine whether end-user or target-system designer predefined
14 multi-part performability requirements are satisfied.

15 The failure scenario generator can be “monolithic”, in the sense that it is designed
16 expressly for the particular kind of target system under analysis, or it can be
17 “synthesized”, in the sense that it is constructed using knowledge of the components and
18 their relationships, and their predetermined failure data. This information can be
19 exploited using any of several well-known mechanisms in the literature of system failure
20 analysis, such as Markov models.

21 One possible optimization is enabled by the introduction of “macro-components”:
22 families of individual components that can be treated as a single entity for failure-
23 scenario analysis.

24 For example, if a disk array uses 3-way mirroring across three identical disk
25 drives, then the failure of any single one of those three disk drives has the same
26 probability of occurrence, and the same impact on performance. This means that the
27 failure-scenario generator 502 may therefore generate a single failure scenario instead of
28 three similar ones, for improved efficiency. More formally, this is equivalent to grouping
29 a set of similar states “ S'_1, \dots, S'_b ” together in the same “ (sc_i, p_i) ” pair, where “ sc_i ” would
30 be any of the grouped states “ S'_1, \dots, S'_b ”, and “ p_i ” would equal “ $OP(S'_1) + \dots +$
31 $OP(S'_b)$ ”, the probability that any of them occurs. This grouping optimization can greatly
32 decrease the number of failure scenarios to be examined; it can be applied when either
33 many states have the same performance, or when the error introduced by grouping is
34 negligible (e.g., when very unlikely states get ignored as a consequence of grouping).
35 Practical systems often have considerable symmetries in their designs; symmetries are a
36 major source of opportunities for grouping. For example, when multiple similar
37 subsystems contain similar data, they can be considered as indistinguishable for multi-
38 part performability evaluation purposes. Thus, the failure-scenario generator may
39 generate all states where a given number of indistinguishable components comprising the
40 subsystems fail at the same time as a single failure-scenario. For the example, the same
41 data may be replicated onto multiple disks in an LU, so the failure of any one of the disks
42 has the same impact on multi-part performability. Or, if the system hardware is
43 recognized to be approximately symmetrical, extrapolation of the data from one failure
44 mode can be employed. Independently of this, each invocation of the generator 502 may
45 return one or more (sc_i, p_i) pairs, depending on implementation decisions.

Failure scenarios may be generated in any order, as long as failure scenarios covering every state are eventually generated. Any failure-scenario generator that satisfies this property, such as one that generates failure scenarios in random or pseudo-random order, can be used in accordance with the present invention. For all but the smallest systems, however, generating the failure scenarios in random order is impractical, because there are too many unimportant failure scenarios. The implementation in a preferred embodiment of the present invention generates failure scenarios in approximate decreasing order of occurrence probability (in which the pairs with the highest value of "p" are generated first). Because of this property, the multi-part performability evaluator 501 considers as few failure scenarios as possible before making a decision on whether the system satisfies the multi-part performability requirements.

The following is a description of a specific implementation for the present invention of a heuristic program that will efficiently generate failure scenarios, each consisting of a single state. Let "FP(c)" denote the "failure probability" of system component "c" — that is, the likelihood that the system component "c" is not operating correctly (equivalently, this is $(1 - A(c))$, where "A(c)" is the availability of component "c" — the fraction of the time that the component is in a correct, functioning state.) That information can be obtained from manufacturer's specifications and from the average time required to repair or replace the component after it fails (a procedure that typically involves human intervention) using well-known techniques in the art. For simplicity, the program generates exactly one (sc, p) pair per invocation, and that pair corresponds to one system state " S_i ".

- (1) Let "D" represent a failure-free system;
- (2) Let " c_1, c_2, \dots, c_{mf} " be components that can fail independently in "D";
- (3) Let "sf" be the number of concurrent failures being considered in the last invocation (initially 0);
- (4) Let "s" be the ordinal number, among the scenarios with exactly "sf" failures, of the scenario returned in the last invocation (initially 0);
- (5) If there exist exactly "s" scenarios with "sf" concurrent failures, then $sf = sf+1; s = 0$;
- (6) If $sf \leq mf$, then $s = s+1$, otherwise exit [no failure scenarios left for consideration];
- (7) choose " a_1, a_2, \dots, a_{sf} " (where $a_i, i=1, \dots, sf$ are different integers between 1 and mf) such that there are exactly " $s-1$ " scenarios with "sf" concurrent failures more likely to occur than " $c_{a1}, c_{a2}, \dots, c_{asf}$ " [i.e., pick the " s^{th} " most likely scenario with "sf" concurrent failures];
- (8) set $sc = D$ with components $c_{a1}, c_{a2}, \dots, c_{asf}$ marked as failed [construct representation of failure-scenario];
- (9) set $p = FP(c_{a1}) \times FP(c_{a2}) \times \dots \times FP(c_{asf}) \times (1 - FP(c_{b1})) \times \dots \times (1 - FP(c_{b(mf-sf)}))$, where $c_{b1}, \dots, c_{b(mf-sf)}$ are all the components that did not fail in "sc" [compute probability for scenario "sc"]; and
- (10) return (sc, p) .

Performance Predictor 503:

1 There exist well-known techniques for implementing a performance predictor
2 503; the techniques used are typically tightly tied to the particular application domain.
3 One type of known manner performance prediction is described in the Lee and Katz
4 article, *supra*.

5 In essence, the performance predictor takes as input, represented by an arrow
6 labeled [5] a failure scenario (such as the sc in the current (sc, p_i) output of 502) and a
7 workload description. The performance predictor returns the predicted performance of
8 the workload under the current input failure scenario, represented by an arrow labeled
9 [6], possibly under a workload that may also be communicated to the performance
10 predictor. Other such known manner performance predictor modules may be adapted and
11 employed in accordance with the present invention.

12 *Multi-part Performability Evaluator 501:*

13 This section details the method for determining whether a multi-part
14 performability requirement can be met by the target system. The techniques presented
15 hereinafter can also be used to allow target-system designers to specify independent
16 performability requirements for independent pieces of data in the input workload.

17 Turning now also to Figure 6, calculations using the multi-part performability
18 evaluator 501 to predict whether a multi-part performability requirement is satisfied in
19 accordance with the present invention are made in conjunction with the failure-scenario
20 generator 502 and performance predictor 503. In general, the calculation is run only until
21 a system-wide PASS-FAIL conclusion can be accurately reported, namely, as quickly as
22 possible.

23 There are two inputs to this process. The first input data 601, represented in
24 Figure 5 by an arrow labeled [1], is a system description for the system configuration
25 being considered. For example, in analyzing the exemplary system in Figure 1, the
26 system description comprises components including controllers, cache memory, busses,
27 arrays of disk drives, and appropriate interface devices. For simplicity of explanation,
28 host computer 111, 111' failures are ignored. Each component has a supplied, predicted
29 failure rate (perhaps supplied by its manufacturer) – typically expressed as the annual
30 failure rate (AFR) or its inverse, the mean time before failure (MTBF).

31 The other input data 603, represented in Figure 5 by an arrow labeled [2], is the
32 target-system designer specified multi-part performability specifications. It may also
33 include workload information.

34 Again using Figure 1 and the previously described example system 100, the
35 predefined multi-part performability specification can include minimal levels of
36 performance that the target system needs to satisfy and the fraction of the time a
37 particular performance level is acceptable, curve 201, Figure 2. In addition, it can
38 include descriptions of the workload demands that will be put on the target system (for
39 example, average request rate in IO/s, mean IO request size, average number of requests
40 that are sequential in LU address space, as described hereinbefore). The workload
41 parameters are preferably encoded in a compact, computer-readable form.

42 Next, the multi-part performability evaluator 501 invokes the failure-scenario
43 generator 502, step 607, represented in Figure 5 by an arrow labeled [3]. The failure-
44 scenario generator 502 returns the first failure scenario, namely the (sc, p_i) pair

1 representative of the approximately most likely failure. Note that more than one (sc,p)
2 pair may be returned.

3 These first-cut failure-scenarios are evaluated one-by-one by the performance
4 predictor 503, step 609. The inputs to step 609, represented in Figure 5 by an arrow
5 labeled [5], are the failure scenario under consideration and the workload parameters
6 (input to the system in the flow represented in Figure 5 by an arrow labeled [2]). For
7 each (sc,p) pair, the performance predictor 503 returns an estimation (represented in
8 Figure 5 by an arrow labeled [6]) of the performance the workload would achieve in the
9 system configuration being considered if the failures (if any) in the current failure
10 scenario "sc" had occurred. Per the exemplary embodiment, the return value may be in
11 the form of the average number of IO/s for the current degraded operational state under
12 consideration.

13 One advantage of the present invention is that availability and performance
14 calculations are decoupled and done independently; first, determining which failure
15 scenarios are relevant to the analysis and how likely they are to occur, and second,
16 computing performance predictions only for the relevant failure scenarios.

17 Next, step 611, the multi-part performability evaluator 501 creates a multi-part
18 performability function (expressed as a curve in Figure 3, 301). That is, the data returned
19 [6] from the performance predictor 503 is converted to a format compatible with
20 comparison to the target-system designer's multi-part performability requirements [2].

21 In Figure 4, for example, the segment 403 of the derived candidate system's
22 multi-part performability curve (analogous to curve 301, Figure 3 and 301', Figure 3A)
23 corresponds to a single failure scenario. By incrementally building the multi-part
24 performability curve of the candidate system configuration, partial results are combined
25 in a way that is accurate and avoids unnecessary computations.

26 The multi-part performability function produced by step 611 and the multi-part
27 performability requirement function in input data 603 are then compared, step 613. In
28 other words, a determination is made as to whether any curve crossing has occurred (see
29 e.g., Figure 3 versus Figure 3A, described above). The following process steps describe
30 on way in which this comparison can be performed; those skilled in the art will recognize
31 that other ways of achieving this comparison are possible:

- 32
- 33 (1) set i = 1;
 - 34 (2) generate the next state " S_i " and its occurrence probability
35 " $OP(S_i)$ ", using a failure-scenario generator 502 (supra);
 - 36 (3) compute " $U(S_i)$ " using a performance predictor module 503
37 (supra);
 - 38 (4) if,
39 i
40 $\sum_{k=1} OP(S_k) I(U(S_k) \geq f_j) \geq f_j$, for all $j=1,2,\dots,n$,
 - 41 then the system satisfies the multi-part performability requirements
42 (in other words no curve crossing can occur; see e.g., FIG. 3), exit and
43 report; or
 - 44 (5) if,

1 i
2 $\sum_{k=1}^i OP(S_k) I(U(S_k) < r_j) \geq 1-f_j$, for any $j=1,2,\dots,n$,
3
4 then the system fails the multi-part performability requirements (in
5 other words a curve crossing has occurred; see e.g., FIG. 3A), exit and
6 report; otherwise,
7 (6) set $i = i + 1$ and go to step (2).
8
9 The results of the comparison, step 613, may be that the analysis is incomplete,
10 namely, the failure scenarios considered so far were insufficient to make a decision, in
11 that case the process loops to the next logical failure scenario for consideration. In other
12 words, it should be recognized that it may take a number of loops of the process through
13 multiple failure-scenarios before computing a multi-part performability function that is
14 sufficient to make a definite decision. If the multi-part performability function is detailed
15 enough to ascertain whether the end user or target-system designer's multi-part
16 performability requirements are satisfied, a report issues that the system passed, 615 (see
17 Figure 3) or failed 617 (see Figure 3A).

18
19 Figure 4 shows an example in which only a portion of the candidate system's
20 multi-part performability curve 403 must be generated before reaching a decision,
21 provided that the relevant failure scenario(s) is examined as early as possible in the
22 process. Since the requirement curve 201 calls for at most 20% downtime and segment
23 403 implies that downtime is 10% or less, it is not necessary to further refine the analysis
24 of area 401 of the graph. Analogously, if it is known (e.g., because of the number of
25 failures) that all failure scenarios with performance levels lower than segment 403 fall
26 into area 401, there is no need to further examine area 402 either. Accordingly,
27 computing the performance prediction for the single failure scenario corresponding to
28 segment 403 is enough to determine that the system satisfies the requirements.
29
30 Depending on the particular system to be analyzed in accordance with the present
31 invention, the performability evaluator 501 may determine whether given multi-part
32 performability requirements are satisfied without evaluating all failure scenarios
33 generated by the failure-scenario generator 503. For example, a particular requirements
34 curve 201 may only require 50 IO/s with 40% availability and 10 IO/s with 80%
35 availability. When enough failure scenarios to account for the 40% availability at 50 IO/s
36 and 80% availability at 10 IO/s or better have been analyzed, additional scenarios do not
37 have to be evaluated; in other words, it is not necessary to spend the additional effort to
38 analyze the behavior of the system for the remaining 20% of the time.

39 Unlike prior solutions, the method of the present invention does not require
40 building a model of an entire solution (although providing an option to do so) and
41 incurring the computational effort and delay of estimating the performance of system
42 states that will not be relevant to the final analysis. The ordering heuristic used in the
43 proposed preferred embodiment of the failure-scenario generator 502 has the goal of
44 analyzing only as many failure scenarios as necessary.

45 In summary, given a multi-component system and a multi-part performability
46 requirement for it, the present invention generates an accurate determination of whether

1 the system will fulfill the multi-part performability requirements. Said determination is
2 rapidly generated by modest computational means.

3 The foregoing description of the preferred embodiment of the present invention
4 has been presented for purposes of illustration and description. It is not intended to be
5 exhaustive or to limit the invention to the precise form or to exemplary embodiments
6 disclosed. Many modifications and variations will be apparent to practitioners skilled in
7 this art. For example, the present invention can be readily adapted to the analysis of other
8 complex system candidates such as for computer-printer networks, telecommunications
9 networks, and the like, or even for fields outside computer-based systems. Similarly, any
10 process steps described might be interchangeable with other steps in order to achieve the
11 same result. The disclosed embodiment was chosen and described in order to best
12 explain the principles of the invention and its best mode practical or preferred
13 application, thereby to enable others skilled in the art to understand the invention for
14 various embodiments and with various modifications as are suited to the particular use or
15 implementation contemplated. It is intended that the scope of the invention be defined by
16 the following claims and their equivalents. Reference to an element in the singular is not
17 intended to mean "one and only one" unless explicitly so stated, but can mean "one or
18 more." Moreover, no element, component, nor method step in the present disclosure is
19 intended to be dedicated to the public regardless of whether the element, component, or
20 method step is explicitly recited in the claims. No claim element herein is to be
21 construed under the provisions of 35 U.S.C. Sec. 112, sixth paragraph, unless the element
22 is expressly recited using the phrase "means for. . .".

23

TOP SECRET - SOURCE CODE